1 Projecting circum-Arctic excess ground ice melt with a sub-grid representation in the Community

2 Land Model

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7 Abstract To address the longstanding underrepresentation of the influences of highly variable ground 8 ice content on the trajectory of permafrost conditions simulated in Earth System Models under a warming 9 climate, we implement a sub-grid representation of excess ground ice within permafrost soils using the 10 latest version of the Community Land Model (CLM5). Based on the original CLM5 tiling hierarchy, we 11 duplicate the natural vegetated landunit by building extra tiles for up to three cryostratigraphies with 12 different amounts of excess ice for each grid cell. For the same total amount of excess ice, introducing 13 sub-grid variability in excess ice contents leads to different excess ice melting rates at the grid level. In 14 addition, there are impacts on permafrost thermal properties and local hydrology with sub-grid 15 representation. We evaluate this new development with single-point simulations at the Lena river delta, 16 Siberia, where three sub-regions with distinctively different excess ice conditions are observed. A triple-17 landunit case accounting for this spatial variability conforms well to previous model studies for the Lena 18 river delta and displays a markedly different dynamics of future excess ice thaw compared to a single-19 landunit case initialized with average excess ice contents. For global simulations, we prescribed a tiling 20 scheme combined with our sub-grid representation to the global permafrost region using presently 21 available circum-Arctic ground ice data. The sub-grid scale excess ice produces significant melting of 22 excess ice under a warming climate and enhances the representation of sub-grid variability of surface 23 subsidence on a global scale. Our model development makes it possible to portray more details on the 24 permafrost degradation trajectory depending on the sub-grid soil thermal regime and excess ice melting, 25 which also shows a strong indication that accounting for excess ice is a prerequisite of a reasonable 26 projection of permafrost thaw. The modeled permafrost degradation with sub-grid excess ice follows the 27 pathway that continuous permafrost transforms into discontinuous permafrost before it disappears, 28 including surface subsidence and talik formation, which are highly permafrost-relevant landscape 29 changes excluded from most land models. Our development of sub-grid representation of excess ice 30 demonstrates a way forward to improve the realism of excess ice melt in global land models, but further 31 developments require substantially improved global observational datasets on both the horizontal and 32 vertical distributions of excess ground ice.

33 1. Introduction

Permafrost soils are often characterized by different types of ground ice that can exceed the pore
space (Brown et al. 1997; Zhang et al., 1999). The presence of such "excess" ground ice can alter the

36 permafrost thermal regime and landscape structure. Widespread thawing of permafrost is expected in a 37 warmer future climate and modeling studies suggest large-scale degradation of near-surface permafrost 38 at the end of the 21st century (Lawrence et al., 2008 & 2011). Melting of ground ice due to active layer 39 thickening releases water in the form of surface runoff, subsurface flow, or both, causing surface 40 subsidence and modifying the local hydrological cycle (West and Plug, 2008; Grosse et al., 2011; Kokelj 41 et al., 2013; Westermann et al., 2016). In addition to containing ground ice, some permafrost soils store 42 massive amounts of carbon, which could be released to the atmosphere in the form of greenhouse gases 43 upon thawing (Walter et al., 2006; Zimov et al., 2006; Schuur et al., 2008), possibly making a positive 44 feedback to amplify future climate change (Koven et al., 2011; Schaefer et al., 2014; Burke et al., 2013). 45 The existence of excess ice and its distribution in permafrost can significantly affect the rate of permafrost 46 thawing (Westermann et al., 2016; Nitzbon et al., 2020), and in turn, the rate of soil carbon release 47 (Hugelius et al., 2014; Schuur et al., 2015; Turetsky et al., 2019). Therefore, better projections of excess 48 ice melt are critical to improve our understanding of the impacts of permafrost thaw on corresponding 49 climatic impacts.

50 Previous studies address excess ice modeling on the local or regional scale, in which the small study 51 area makes it possible for detailed configurations of the cryostratigraphy of permafrost and excess ice 52 based on observations. Simulations for the Lena river delta have retrieved the permafrost thermal 53 dynamics fairly close to the observations with excess ice incorporated in the modeling (Westermann et 54 al., 2016). A two-tile approach allowing lateral heat exchange between two land elements demonstrated 55 that maintaining thermokarst ponds requires the heat loss from water to the surrounding land (Langer et 56 al., 2016). A similar tiling approach has been applied to projecting the landscape changes due to 57 permafrost thaw for ice-wedge polygons and peat plateaus with different features of ice melting and 58 surface subsidence (Aas et al., 2019; Nitzbon et al., 2019).

59 On the global scale, the land components of Earth System Models (ESMs) have significant 60 capabilities of representing key permafrost physics. In the Community Land Model (CLM), for example, 61 the representation of permafrost-associated processes has been continuously improved. By including key 62 thermal and hydrological processes of permafrost, the CLM version 4 (CLM4) has reasonably 63 reproduced the global distribution of permafrost (Lawrence et al., 2008; Lawrence et al., 2012; Slater 64 and Lawrence, 2013). Projections based on the CLM4 under its highest warming scenario (RCP8.5) have 65 shown over 50% degradation of near-surface permafrost by 2100 (Lawrence et al., 2012). Moreover, the 66 recently released CLM5 has more advanced representations of many biogeophysical and biogeochemical 67 processes (Lawrence et al., 2019). A refined soil profile and upgraded snow accumulation and 68 densification scheme in the CLM5 could contribute to simulating more realistic permafrost thermal 69 regimes, whereas upgrades on biogeochemistry improve simulations of soil carbon release in response 70 to permafrost thaw. In addition, an excess ice physics scheme has been implemented in CLM4.5 71 (CLM4.5 EXICE) by Lee et al. (2014), which allowed for the first-order simulation of surface 72 subsidence globally by modeling excess ice melt under a warming climate.

73 The homogeneous distribution of excess ice throughout the grid cell in CLM4.5 EXICE (Lee et al., 74 2014) could cause biases in thaw trajectories in the warming climate. In nature, excess ice forms in a 75 highly localized manner due to a variety of accumulation processes. For instance, segregated ice formed 76 during frost heave differs substantially in excess ice morphology from ice wedges that are formed from 77 repeated frost cracking and freezing of penetrating water. Field measurements illustrate that the depth 78 distribution of ground ice can vary substantially on the order to 10-50 metres horizontally and 0-10 metres 79 vertically (Pascale et al., 2008; Fritz et al., 2011). The horizontal grid spacing of ESMs, on the other hand, 80 usually ranges from one to two degrees (~100-200 km horizontal scale), which makes it impossible to 81 represent localized excess ice. The mismatch in spatial scale between model and the real world raises 82 concerns for the reliability of excess ice modeling in ESMs. Aside from the homogenously initialized 83 excess ice in the grid cell, CLM4.5 EXICE initializes excess ice in the same soil depths globally (below 84 1m), regardless of the varying active layer thickness in circum-Arctic permafrost areas (Lee et al., 2014). 85 Such deficiencies in excess ice parameterization hamper global projections of permafrost thaw including 86 excess ice with ESMs.

87 To narrow the gap between the high spatial variability of excess ice and the coarse grid spacing in 88 the ESMs, we applied a sub-grid approach in representing excess ice in permafrost soils within the CLM5 89 to investigate how presence and melting of excess ice affect land surface physics under a warming climate. 90 We conducted idealized single-point simulations to examine the robustness of model development, and 91 furthermore conducted global simulations using a first-order estimate for the spatial distribution of excess 92 ice and associated cryostratigraphies. Due to the lack of information in global excess ice conditions, it is 93 not the aim of this study to accurately project excess ice melt and surface subsidence in the 21st century, 94 but rather to develop and present a functionable process within a land surface model that can eventually 95 bring permafrost thaw modeling towards a higher degree of accuracy on a global scale. The CLM5 with 96 sub-grid excess ice representation developed through this study would be ready to serve as a proper 97 simulation tool on further advancing global excess ice modeling once new datasets become available.

98 2. Methodology

99 2.1 Sub-grid representation of excess ice in the CLM5

100 The CLM5 model utilizes a three-level tiling hierarchy to represent sub-grid heterogeneity of 101 landscapes, which are (from top to bottom) landunits, columns, and patches (Lawrence et al., 2019). 102 There is only one column (the natural soil column) that is under the natural vegetated landunit, which 103 represents soil including permafrost. In this study, we modify the CLM5 tiling hierarchy by duplicating 104 the natural vegetated landunit, making extra landunits for prescribing up to three different excess ice 105 conditions in permafrost (Figure 1). The original natural vegetated landunit is considered as "natural 106 vegetated with no excess ice" (hereafter no ice landunit), while we denote the additional landunits as 107 "natural vegetated with low content of excess ice" (hereafter the low ice landunit), "natural vegetated 108 with medium content of excess ice" (hereafter the mid ice landunit), and "natural vegetated with high 109 content of excess ice" (hereafter the high ice landunit). The sub-grid initial conditions of excess ice are

imported as part of the surface data, which includes the variables of volumetric excess ice contents,depths of the top and bottom soil layer of added excess ice, and the area weights of the four landunits.

112 We adopted the excess ice physics from CLM4.5_EXICE (Lee et al., 2014), including 113 thermodynamic and hydrological processes. The added excess ice is evenly distributed within each soil 114 layer. Whereas the original CLM5 model already represents the dynamics of pore ice, our representation 115 of excess ice physics only addresses the ground ice bodies that exceed soil pore space. The volumetric 116 excess ice content in this study is defined as the ratio of the volume of excess ice in a soil layer to the 117 volume of the whole soil layer. For example, a 50% volumetric content of excess ice means the excess 118 ice body occupies 50% volume of a soil layer, while the rest of soil (and pore ice) occupies the other 50% 119 volume of the soil layer. If not otherwise notified, the parameter of volumetric ice content in this 120 manuscript refers only to that of excess ice bodies. After adding excess ice, the soil layer thickness 121 increases accordingly. Because ice density is considered constant, the increase of soil layer thickness is 122 linearly proportional to the volumetric content of excess ice. For example, adding an excess ice body 123 with a 50% volumetric excess ice content doubles the soil layer thickness of the corresponding soil layer. 124 The revised algorithm for thermal conductivity and heat capacity of soil involves the effects of added 125 excess ice, while the revised phase change energy equation allows excess ice to melt. The meltwater adds 126 to soil liquid water in the same soil layer, and it can move to the above layer if the original layer is 127 saturated. Such numerical implementation replicates how the melt excess ice eventually converts to 128 runoff and discharges from the soil in case of well-drained conditions. As excess ice melts, soil layer 129 thickness decreases, which corresponds to surface subsidence due to excess ice melt. In our model 130 parameterization, excess ice only melts and does not re-form since the applied excess ice physics does 131 not account for the different ice formation processes.

132 Aside from sub-grid tiles for excess ice, we acknowledge that the version upgrade from CLM4.5 to 133 CLM5 as the base model modifies the results of excess ice melt compared to the results from Lee et al. 134 (2014). By default, CLM5 represents soil with a 25-layer profile, for which the top 20 hydrologically 135 active layers cover 8.5 metres of soil. There are additional 10 soil layers and it is 4.7 metres deeper 136 compared to the default hydrologically active soil layer profile in CLM4.5, not to mention the 137 substantially more complex biogeophysical processes (Lawrence et al., 2019). Therefore, we developed 138 the sub-grid representation of excess ice within the framework of the latest version of CLM. The 139 duplicated landunits prolong computation time by roughly 10% compared to the original CLM5. We are, 140 therefore, confident that our model development is highly efficient in addressing the sub-grid excess ice 141 and subsequent permafrost thaw.

We examined the sensitivity of sub-grid excess ice initialization by conducting idealized experiments (see supplemental material). For overall<u>Overall, for</u> the same amount of excess ice in one grid cell located in the same depth, a higher volumetric excess ice content along with a smaller area weight results in a later start of excess ice melt and a <u>smaller-lower</u> melting rate. The different melting features from different sub-grid distribution<u>s</u> of excess ice then leads to different hydrological impacts to the permafrost soil. The results of the idealized experiments in this way verify the necessity of involving introducing sub-grid configuration of excess ice to the CLM that is with has a typical horizontal
 grid spacing of 1-2 degrees. More details are available in the supplemental material.

150 2.2 Single-point simulations for the Lena river delta, Siberia

We conduct single-point simulations for the Lena River delta and compare the CLM5 model results 151 152 to reference simulations with the CryoGrid3 model for the same location (Westermann et al., 2016). 153 Abundant background information is available on the soil and ground ice dynamics from both 154 observation and modeling, making the Lena river delta a suitable location to further evaluate our model 155 development. The Lena river delta can be broadly categorized into three different geomorphological units 156 that have distinctively different subsurface cryostratigraphies of excess ice (Schneider et al., 2009; Ulrich 157 et al., 2009). In the eastern and central part of the river delta, ground ice has been accumulated in the 158 comparatively warm Holocene climate. The subsurface sediments (hereafter denoted as "Holocene 159 ground ice terrain") are generally super-saturated with wedge ice that can extend up to 9 metres underground with the volumetric contents of total ground ice (pore ice + excess ice) ranging from 60-160 161 80% (Schwamborn et al., 2002; Langer et al., 2013). On the other hand, higher excess ice contents are 162 found in Pleistocene sediments in the Lena River Delta (hereafter the "Yedoma Ice complex"), which 163 are characterized by Yedoma type ground ice (Schirrmeister et al., 2013), which can reach depths of up 164 to 20-25 metres deep and volumetric contents of total ground ice as high as 90% (Schwamborn et al., 165 2002; Schirrmeister et al., 2003 and 2011). Finally, the Northwestern part of the delta features sandy 166 sediments and is characterized by low excess ice contents (hereafter denoted the "no excess ice terrain"; 167 Rachold and Grigoriev, 1999; Schwamborn et al., 2002).

168 We determine the area weights of excess ice landunits in one single point based on the spatial pattern 169 of three subregions (Fedorova et al., 2015). The cryostratigraphy and the volumetric contents of excess 170 ice strictly follow those in Westermann et al. (2016). Noting that the excess ice initialization scenario in 171 Westermann et al. (2016) does not necessarily represent the realistic excess ice condition for the Lena 172 river delta, the purpose of applying the same excess ice cryostratigraphy as in Westermman et al. (2016) 173 is to evaluate our model development by addressing intercomparisons between model results. Meanwhile, 174 we did not customize soil properties for different landunits as in Westermann et al. (2016), as our model 175 development does not support varying soil properties for different sub-grid landunits. We also directly 176 apply the snow accumulation physics in the CLM rather than customizing the snow density. By default, 177 the current model does not form thermokarst lakes as the meltwater from excess ice melt becomes surface 178 runoff and is removed from the grid cell. To apply the sub-grid representation, we initialize the case with 179 three landunits (the triple-landunit case) that respectively represent the three terraces in the Lena river 180 delta. We also initialize an "average ice single-landunit" case without the sub-grid representation of 181 excess ice. The excess ice amount for each soil layer in the average ice single-landunit case is initially 182 the same as that in the triple-landunit case. The volumetric content of excess ice is determined by spatial 183 averaging those for three excess ice landunits in the triple-landunit case. Detailed information on the 184 applied excess ice conditions for both cases is listed in Table 1.

185 We employed the single-point forcing data from in Westermann et al. (2016) for the Lena river delta from 1901 to 2100, which is based on the CRU-NCEP (http://dods.extra.cea.fr/data/p529viov/cruncep/) 186 187 data set for the historical period (1901-2005) and the CCSM4 model output under the RCP4.5 scenario 188 for the projected period (2006-2100), but downscaled with in-situ observations. We run 100-year spin-189 up simulations in order to stabilize the permafrost thermal regime after adding excess ice. Spin-up 190 simulations are produced by running the model with cycled 1901-1920 climatological data. The purpose 191 of spin-up simulations is to stabilize ground temperatures and volumes of excess ice bodies. The 100-192 year length for spin-up is sufficient, as the model is run in Satellite Phenology (SP) mode that does not 193 involve slowly evolving biogeochemical processes such as soil carbon accumulation. Moreover, we 194 address idealized single-point simulations for additional permafrost locations with both continental and 195 maritime climate that showcase the difference to Lee et al. (2014), the results of which are included in 196 the Supplementary material.

197 2.3 Global simulations of excess ice melt

198 The information available for the spatial distribution of excess ice and associated cryostratigraphies 199 on the global scale is generally not as detailed as in the Lena river delta due to the lack of observations. 200 For our global simulations we employ the widely used "Circum-Arctic Map of Permafrost and Ground-201 Ice Conditions" (hereafter the CAPS data; Brown et al., 2002) as data source, while we translate the 202 ground ice condition in the CAPS data to different excess ice stratigraphies as model input data. The 203 CAPS permafrost map categorizes the global permafrost area into classes coded by three factors (i) 204 permafrost extent (c = continuous, d = discontinuous, s = sporadic, and i = isolated), (ii) visible ground 205 ice content (h = high, m = medium, and l = low), and (iii) terrain and overburden (f = lowlands, highlands,206 and intra- and intermontane depressions characterized by thick overburden cover, and r = mountains, 207 highlands ridges, and plateaus characterized by thin overburden cover and exposed bedrock), resulting 208 in more than 20 different varieties in permafrost characteristics (Figure 2). For the simulations, we only 209 use the CAPS distinction between the three classes: high, medium and low ice contents. We qualitatively 210 categorize excess ice types with typical cryostratigraphies for which observations are available, 211 recognizing that this is a crude first-guess of the global distribution of ground ice which needs to be be 212 improved in future studies.

213 The high ice CAPS classes (e.g. chf, chr, and dhf) in central and eastern Siberia, as well as in Alaska, 214 partly coincide with Yedoma regions (Kanevskiy et al., 2011; Grosse et al., 2013). The cryostratigraphy 215 of the high ice landunit is therefore broadly oriented at the excess ice contents and distribution in intact 216 Yedoma, which is characterized by massive ice wedges leading to typical average volumetric content of 217 total ground ice in the range from 60% to 90% (Schwamborn et al., 2002; Kanevskiy et al., 2011). We 218 therefore set the volumetric content of excess ice in the high ice landunit to 70%, and we put excess ice 219 in all the soil layers between 0.2 metres below the active layer and the bottom of hydrologically active 220 soil layer (8.5 metres). The onset depth of the excess ice just below the active layer is based on the 221 assumption of active ice aggradation which occurs at or below the permafrost table, e.g. the formation of 222 wedge or segregation ice. Initializing high excess ice content throughout the whole soil layer imitates the

223 cryostratigraphy of Yedoma type ice, while roughly 65% of the high ice landunit is located out of the 224 observed Yedoma regions (Schuur et al., 2015). The effects, limitations, and potential improvements of 225 this initialization scenario will be mentioned in the discussion section. For the low ice landunit, we 226 assume both a significantly lower volumetric excess ice content and a smaller vertical extent of the excess 227 ice body. The volumetric excess ice content is set to 25%, and we add excess ice at soil layers within 0.2 228 to 1.2 metres below the active layer, which in particular represents sediments with segregated ice (e.g. 229 Cable et al., 2018), but also accounts for a wide range of different excess ice conditions found throughout 230 the permafrost domain. For the mid ice landunit, we set the volumetric excess ice content to 45% and 231 put excess ice within 0.2 to 2.2 metres below the active layer, making the volumetric excess ice content 232 and vertical extent of which in between those for the low and high ice landunits. The cryostratigraphies 233 determine that excess ice melt in the low ice landunit can result in a maximum of 0.36 metres of surface 234 subsidence, while excess ice melt in the medium ice landunit can result in a maximum of 1.78 m of 235 surface subsidence. For the high ice landunit, the surface subsidence can be more than 10 metres if all 236 excess ice melts, which is expected to vary in space because of the different active layer thickness. For 237 all three landunits, the active layer thickness is determined by the soil temperature profile by the end of 238 the spinup in a no ice case, which is the simulation by the original CLM5 model without excess ice 239 incorporated. Non-permafrost regions in the CAPS data are assigned the no ice landunit for 100% of 240 their area. We emphasize that the prescribed cryostratigraphies are a first-order approximation that can 241 by no means represent the wide variety of true ground ice conditions found in the permafrost domain. 242 Nevertheless, this makes it possible to gauge the effect of excess ice melt on future projections of the 243 permafrost thermal regime, when compared to "traditional" reference simulations without excess ice.

244 We design a tiling scheme prescribing the assignment of landunits for each CAPS class based on 245 previous observations and empirical estimates (Table 2). All CAPS classes in this study are categorized 246 into three levels of volumetric ice content (5%, 15%, and 25%) that are converted from the ranges (<10%, 247 10-20%, and >20%) in the original CAPS data. The goal of our tiling scheme is to determine a 248 combination of area weights of three excess ice landunits for each CAPS class, making the spatially 249 averaged volumetric content of excess ice the same as that for the CAPS class. We assume that all CAPS 250 classes have the same area fraction (20%) of the low ice landunit, and the CAPS classes with a higher 251 ice content are due to the existence of the landunits with a higher content excess ice. We make this 252 assumption based on previous studies that the segregated ice is widely distributed in permafrost. 253 Observational studies have found segregated ice bodies in various continuous permafrost regions across 254 the circum-arctic including West Central Alaska (Kanevskiy et al., 2014), Nunavik, Canada (Calmels 255 and Allard, 2008), and Svalbard (Cable et al., 2018). In discontinuous permafrost regions, segregated ice 256 bodies also commonly exist underneath Palsas and Lithasas, including Fennoscandia (Seppälä, 2011), 257 Altai and Sayan, Russia (Iwanhana et al., 2012), Himalayas (Wünnemann et al., 2008), and Mongolia 258 (Sharkhuu et al., 1999). The volumetric content of visible segregated ice bodies mentioned above ranges 259 widely from 10-50% (Gilbert et al., 2016).

Given the tiling scheme prescribed above, all CAPS classes are assigned a 20% area of low ice
 landunit. Correspondingly, the CAPS classes with 15% volumetric ice content are assigned another 14%

262 area weight for mid ice landunit on top of the CAPS classes with 5% volumetric ice content, while the 263 CAPS classes with 25% volumetric ice are assigned another 22% area for high ice landunit on top of the 264 CAPS classes with 15% volumetric ice content. The classes of "chf" and "chr" are the exceptions as their 265 corresponding regions are typically with the landscape of Yedoma or ice wedge polygonal tundra or both 266 (Kanevskiy et al., 2011; Gross et al., 2013). We therefore assign only the low and high ice landunits for 267 these two CAPS classes. Summing up the landunit fractions for all the CAPS grid cells within each CLM 268 grid cell obtains the area weights on the grid level that are stored in the surface data file. Figure 3 shows 269 a schematic plot for the initialization scenario and the area covered by different excess ice landunits as 270 the result of sub-grid excess ice initialization in the global simulation case. Note that excess ice for some 271 regions (e.g. Southern Norway and the Alps) can completely melt out during the spinup period since the 272 CLM initial condition prescribes overly warm (non-permafrost) soil temperature for these regions.

273 In this study, we define the grid cells or landunits with permafrost as the ones having at least one 274 hydrologically active soil layer that has been frozen in the last consecutive 24 months. In this case, we 275 define fully degraded permafrost when all landunits in one grid cell have an active layer thickness of 276 more than 6.5 metres, recognizing that in reality permafrost at many localities may continue to exist at 277 to greater depths. We also prepare a "grid-average ice case" by applying the same total amount of excess 278 ice as in the sub-grid ice case in each soil layer, but using only one landunit instead of three that account 279 for the sub-grid variability of excess ice. The volumetric content of excess ice in the single landunit is 280 calculated as the spatial average of those in the three landunits in the triple-landunit case. This grid-281 average ice case provides a reference to evaluate the effects of the sub-grid excess ice representation on 282 the global scale. Finally, we simulate a reference case without excess ice, denoted the "no ice case" in 283 the following. Details on the three cases for the global simulations are listed in Table 3. All global cases 284 are forced by the 3rd version of Global Soil Wetness Project forcing data (GSWP3; Kim et al., 2012), 285 running in the Satellite Phenology (SP) mode. The International Land Atmosphere Model Benchmarking 286 (ILAMB; Collier et al., 2018) project has indicated the superior performance of GSWP3 data forcing the 287 CLM5 in the SP-only mode (http://webext.cgd.ucar.edu/I20TR/ build 090817 CLM50SPONLY CRUNCEP GSWP3 WFDEI/in 288 289 dex.html). We conducted a 100-year spin-up using the 1901-1920 climatology before conducting 290 historical period simulations covering 1901-2005. The anomaly forcing under the RCP8.5 scenario on 291 top of the 1982-2005 climatology forces simulations in the projected period.

292 3. Result<u>s</u>

293 3.1 Excess ice melt simulations for Lena River delta cryostratigraphies

By the end of the spinup in the triple-landunit case, the active layer thickness is 0.85 m, 0.55 m, and 0.45 m for the ice-poor terrain, the Holocene ice wedge terrain, and the Yedoma ice complex, respectively. On the other hand, the active layer thickness for the average ice single-landunit case is 0.85 m, which is the same as in the no excess ice terrain in the triple-landunit case. For the average ice single-landunit case, a small amount of excess ice (24 kg m⁻²) melts during the spinup period, resulting in 2.6 cm surface subsidence throughout the grid. 300 For the Yedoma ice complex, very little excess ice melt in the 1950s, and it stabilizes afterwards 301 until the late 2000s when substantial ice melt and surface subsidence starts to occur. For the Holocene 302 ground ice terrain, there is no excess ice melt before the late 2010s. By the year 2100, the Yedoma ice 303 complex has exhibited nearly 4 metres of surface subsidence, while the Holocene ground ice terrain has 304 about 0.6 metres of surface subsidence (Figure 4). For the average ice single-landunit case, the noticeable 305 excess ice melt and surface subsidence starts in the late 2010s, which creates about 0.5 metres of surface 306 subsidence by 2100. The magnitude of surface subsidence in the average ice single-landunit case is lower 307 than both the Holocene ground ice terrain and the Yedoma ice complex in the triple-landunit case.

308 On the grid scale, the total excess ice melt is higher in the average ice single-landunit case than in 309 the triple-landunit case (Figure 5). By the year 2100, the average ice single-landunit case has about 30 310 kg m⁻² more excess ice melt than the triple-landunit case. The difference in excess ice on the grid level 311 results from the different volumetric content of excess ice caused by the spatial averaging. In this way, 312 the sub-grid representation of excess ice can potentially also provide more detailed and realistic 313 representation of model variables on the grid level. This is particularly important for the CLM5, which 314 serves as the land component in Earth System Models, which requires the coupling between interacting 315 components on the grid level.

316 Compared to Westermann et al. (2016), the CLM5 with sub-grid excess ice simulates slightly less 317 ($\sim 20\%$ less) surface subsidence by 2100 for both the central delta and ice complex. We consider this a 318 good agreement as we do not expect a closer fit of the model results due to substantial differences in the 319 model physics (for example, the Cryogrid3 simulations in Westermann et al. (2106) lack a representation 320 of the subsurface water cycle). What is in common between these two studies is the earlier start of excess 321 ice melt and more surface subsidence in the ice complex than in the central delta. The CLM5 with sub-322 grid excess ice also exhibits the varying active layer thickness with different excess ice conditions as 323 Cryogrid3 does. These results suggest that the new model development enables small-scale variability in 324 excess ice melt and subsequent impacts in agreement with previously published modeling efforts.

325 3.2 Global projection of permafrost thaw and excess ice melt

326 Single-point simulations have shown that the varying excess ice cryostratigraphies for different 327 landunits result in sub-grid variabilities of excess ice melt and surface subsidence under the warming 328 climate. The same features remain in the sub-grid ice case within the global simulations that excess ice 329 in the low ice landunit can completely melt out throughout the circum-Arctic permafrost region by the 330 end of the 21^{st} century (Figure 6). The modeled magnitude of surface subsidence is similar to the ~10 cm 331 surface subsidence observed in Barrow and West Dock in the early 21st century (Shiklomanov et al., 2013; Streleskiy et al., 2017). The magnitude of surface subsidence is also comparable to the 1-4 cm 332 333 decade⁻¹ surface subsidence rate on average over the North Slope of Alaska observed by satellite 334 measurements since the 1990s (Liu et al., 2010). In comparison, the absence of surface subsidence for 335 Arctic Alaska modeled by Lee et al. (2014) is due to an overly deep (1 m deep) excess ice initialization 336 depth. By the year 2100, most ice in the medium ice landunit melts away in the sub-arctic region, while 337 there is less ice melt in the colder regions such as the North Slope of Alaska and the central Siberia. The

high ice landunit has the greatest surface subsidence among the three because of its high excess icecontent, leading to 2-5 metres of surface subsidence by the year 2100.

340 The existence of excess ice modulates the thermal regime of permafrost soil and is a major control 341 on permafrost degradation trajectories in a warming climate. Permafrost with excess ice consistently 342 exhibits delayed permafrost degradation compared to the no ice case (Figure 7). For the no ice case 343 modeled by the original CLM5, more than half of the permafrost area undergoes degradation by the end 344 of the 21st century. By 2100, the only areas where permafrost remains are the North Slope of Alaska, 345 Northern Canada, and the majority of the land area in Northern Siberia. The areas with remaining 346 permafrost in the year 2100 under the RCP8.5 scenarios are substantially larger compared to the CLM4 347 simulations, in which nearly all permafrost in Eurasia becomes degraded (Lawrence et al., 2012). For the 348 grid-average ice case, the presence of excess ice stabilizes the permafrost thermal regime and thus 349 sustains a larger permafrost area on a global scale in the simulation. For example, permafrost areas in 350 some subarctic regions in the eastern and western Siberia, as well as part of the Arctic coastal regions in 351 Yukon Territory, Canada, remain in the grid-average ice case by 2100. Compared to the grid-average ice 352 case, even more permafrost areas are sustained in the sub-grid ice case, most of which are located in 353 southern Siberia. In the subarctic regions in Alaska and Northwest Canada as well as part of the central 354 Siberia, permafrost degradation is delayed from the 2040s in the grid ice case to the 2080s in the sub-355 grid ice case. We emphasize that permafrost is only sustained according to the accepted temperature-356 based definition (ground material at temperature below zero for two consecutive years), but excess ice 357 continuously melts in this process, which energetically is a different mode of permafrost degradation, 358 similar to a negative mass balance of glaciers and ice sheets.

359 In the sub-grid ice case, the landunits with high excess ice contents lead to more grid cells for which 360 permafrost conditions remain in the year 2100 compared to the grid-average ice case. On the other hand, 361 permafrost with excess ice only covers a fraction of a grid cell. Among the permafrost degradation 362 trajectories in the three global simulation cases (Figure 8), the sub-grid ice case can provide a more 363 detailed picture on the timing of permafrost degradation. Grid cells become 'partially degraded 364 permafrost' if landunits with excess ice still contain permafrost, which phenomenologically is a more 365 realistic representation that also makes it possible to represent the permafrost distribution in the 366 discontinuous and sporadic permafrost zones. On the other hand, only "fully degraded permafrost" and 367 "remaining permafrost" can be distinguished for the no ice and grid-average ice case. Under the warming 368 climate in the 21st century, the existence of excess ice, especially the high content of excess ice, has a 369 stabilizing effect on soil temperature that delay the disappearance of permafrost on the sub-grid level. 370 Therefore, by the year 2100, there are regions with partially degraded permafrost in between intact and 371 degraded permafrost (Figure 8). For example, in western Siberia, the Pacific coastal area of eastern 372 Siberia, Northwestern Canada, and along the Brooks Range in Alaska, taliks form for landunits with low 373 excess ice contents which leads to partially degraded permafrost regions. Therefore, permafrost 374 degradation exhibits a gradual transition from continuous to discontinuous permafrost, and to non-375 permafrost regions. Some of these regions also encounter substantial surface subsidence in the high ice 376 landunit (> 5 m) (Figure 6).

We further compare the total permafrost area (defined as landunits with active layer thickness < 6.5 metres) in the three cases throughout time. The differences in permafrost area increase from the gridaverage ice case and sub-grid ice case to the no ice case at a rate of 1000 km² per year until 2050 (Figure 9). After 2050, the area difference of permafrost in the grid-average ice case and no ice cases rapidly increases, which reaches nearly one million km² by 2100. In the sub-grid ice case, the rate of increase remains relatively unchanged after 2050, resulting in an about 0.2 million km² larger permafrost area than that in the no ice case.

384 4. Discussion

385 The aim of the sub-grid excess ice representation in the CLM5 is to facilitate long-term global 386 projection of excess ice melt and surface subsidence in the permafrost regions. Results from our idealized 387 sensitivity experiments (see supplemental material) impliesy that overly low volumetric content of excess 388 ice, such as the grid-average ice case in this study and that in Lee et al. (2014), result inproducing an 389 overly early start of excess ice melt and an overly high melting rate. This It is because the a higher content 390 of excess ice covering a smaller area takes longer to absorb enough heat from the atmosphere to satisfy 391 the latent heat of fusion requirements and absorb enough latent heat of fusion from the atmosphere before 392 it can start melting. <u>Consequently, a</u> good model performance in this way relies not only on the updated 393 sub-grid representation of excess ice in the global land model, but also on retrieving accurate initial 394 conditions of excess ice. However, -the corresponding observational data for both background excess ice 395 conditions and model evaluation is sparse, considering especially that drastic excess ice melt as modeled 396 until 2100 is only observed in few locations today (e.g. Günther et al., 2015). In the following, we discuss 397 the challenges and limitations of the sub-grid excess ice framework, and how this sub-grid representation 398 can potentially help the development of other CLM components. Both single-point and global test 399 simulations in this study have shown that excess ice melts under a warming climate is sensitive to its 400 initialization depth. The active-layer-dependent excess ice initialization in this study in the global 401 simulation (sub-grid excess ice case) yields excess ice melt and surface subsidence rates in the early 402 2000s that are comparable to observations. The lower depths of the assumed excess ice body control the 403 termination of excess ice melt which at the same time determines the onset of talik formation in many 404 permafrost areas. Due to the scarcity of observational data, it is unclear to what extent the 405 cryostratigraphies assumed in our tiling scheme can reproduce the true vertical extent of excess ice bodies 406 at least in a statistical sense. Even so, we manage to make the prescribed excess ice condition as close to 407 the previous results as possible. Firstly, our tiling scheme on the large scale strictly follows the CAPS 408 data (Brown et al., 2002) in terms of the volumetric excess ice content. Furthermore, statistics by Zhang 409 et al. (2000) suggest the ranges of the vertical extent of ice-rich permafrost of 0-2 metres and 2-4 metres 410 respectively for the CAPS classes with low (5%) and medium (15%) ice content. Comparatively, the 411 vertical extents permafrost with excess ice prescribed by our tiling scheme are respectively 1.36 metres 412 and 3.78 metres for the same CAPS classes, both of which lie within the ranges in Zhang et al. (2000). 413 The vertical extent of ice-rich permafrost for the high ice landunit is much higher than that (4-6 metres) 414 in Zhang et al. (2000), but the unmelted part of the ice bodies does not strongly affect the overall rate of 415 excess ice melt, although the remaining ice can slightly change soil temperature and moisture of the

surrounding permafrost. We therefore imply that our high ice landunit initialization would not induce astrong bias in excess ice melt projection in the 21st century.

Due to the lack of excess ice datasets and observational evidence, our projections of excess ice melt 418 419 and surface subsidence likely have biases that arise from the need to make empirical estimates and 420 simplifications for the excess ice initialization scenarios in the global simulation cases. For example, as 421 the CAPS data is mostly based on visible ice bodies (i.e., not pore ice) (Heginbottom et al., 1995), we 422 used the reported volumetric ground ice content in the CAPS data to approximate the volumetric content 423 of excess ice during model initialization. Ffurther, the determination of volumetric contents of excess ice 424 for three landunits also results from sparse observations and empirical estimates. The prescribed excess 425 ice cryostratigraphies ignore ice morphology and the variation of volumetric content of excess ice with 426 soil depth, regarding excess ice as homogeneous within each assigned sub-grid ice content type (low, 427 mid, or high) (Figure 3, upper panel).. For the high ice landunit, we simplify the cryostratigraphy 428 initialization to Yedoma type ice, which prescribes overly thick excess ice bodies out of the Yedoma 429 regions (Schurr et al., 2015). A deficiency in the current version of source code prevents us from 430 initializing non-Yedoma wedged ice for the high ice landunit where it occurs outside- of the Yedoma 431 region. Future versions of our model development will have more freedom in the stratigraphic 432 configuration of excess ice, which will make it possible to prescribe different cryostratigraphies of the 433 same landunit (e.g. the high ice landunit) for different locations. Because of the above shortcomings in 434 the excess ice initialization, we do not expect the modeled excess ice melt in this study to be an adequate 435 representation of reality. DHowever, direct ingestion of new or improved observational improved 436 observational data sets of excess ice contents and cyostratigraphies would likely yield more accurate 437 results, however. However, a spatially distributed global dataset with quantitative information on excess 438 ice stratigraphies does not exist at present. We emphasize that for a better projection of excess ice melt, 439 more observational data of excess ice distribution and surface subsidence is required to further evaluate 440 and validate the new model implementation of excess ice. On the regional scale, Jorgenson et al. (2008) 441 presented a permafrost map of total ground ice volume for the uppermost 5 metres of permafrost based 442 on both observations and estimates for Alaska. In addition, O'Neill et al. (2019) compiled permafrost 443 maps for Northern Canada by paleographic modeling, mapping the abundances of three types of excess 444 ice respectively. Further improvements of model results depend on additional observationally 445 constrained datasets of excess ice conditions on the global scale.

446 The area weights of the excess ice landunits (Table 2) in the global simulation are obtained from the 447 higher-resolution CAPS points located within a CLM grid cell. However, complex landscape 448 development, such as thermokarst ponds, requires knowledge of the metre-scale distribution, for example 449 the extent and geometry of individual ice wedges (Langer et al., 2016; Nitzbon et al., 2019), which cannot 450 be represented with the still coarse-scale excess ice classes from the CAPS map. One possible solution 451 to represent this could be to include another layer of sub-grid tiles below the CLM landunit level, where 452 the individual tiles can interact laterally. This would allow for the representation of small-scale 453 permafrost features within a large-scale landunit with a given excess ice content. An example of how this 454 could work is given by Aas et al. (2019) who simulated both polygonal tundra and peat plateaus with a

455 two-tile interactive setup. This is also similar to the recent representation of hillslope hydrology by 456 Swenson et al. (2019), where sub-grid tiles (on the column level in CLM) were used to represent different 457 elements in a representative hillslope. In the future development of CLM, this could be part of a more 458 generic tiling system where lateral heat and mass fluxes could be switched on and off to represent a wide 459 range of land surface processes that are currently ignored or parameterized in LSMs. Fisher and Koven 460 (2020) have discussed the challenges and opportunities in such an adaptive and generic tiling system. 461 We would also advocate for enhancing current tiling schemes in such a direction, which could 462 substantially improve the realism in the representation of permafrost landscapes in LSMs. However, the 463 success of such a tiling approach will rely heavily on the availability of adequate observational data, 464 further highlighting the need for observational efforts and close collaboration between field scientists 465 and modelers.

466 The more detailed simulation of permafrost degradation trajectory with a sub-grid representation of 467 excess ice also builds more potential on better modeling the permafrost-carbon feedback with 468 biogeochemistry activated (CLM5BGC). Excess ice stabilizes the permafrost thermal regime, therefore 469 alter the rate of carbon releasing from the permafrost (Shuur et al., 2008). Improved projections of 470 permafrost warming could also enhance modeling of vegetation type changes (e.g. shrub expansion) that 471 determines the nitrogen uptake to the atmosphere (Loranty and Goetz, 2012). On the other hand, the 472 possibility to simulate surface subsidence and excess ice meltwater formation also opens the possibility 473 of a more accurate representation of wetland formation. The increase in the area of wetland and soil 474 moisture have an impact of the balance of CH₄ and CO₂ releasing from the permafrost as more organic 475 matter could decompose in an anaerobic pathway (Lawrence et al., 2015; Treat et al., 2015). Compared 476 to the parameterized inundated area simulation in the CLM5 (Ekici et al., 2019), a process-based wetland 477 physics scheme together with the sub-grid representation of excess ice in this study would substantially 478 contribute to the biogeochemical modeling over the circum-arctic area.

479 5. Conclusion

This study develops a sub-grid representation of excess ice in the CLM5 and examines the impacts of the existence and melting of excess ice in the sub-grid scale in a warming climate. Extra landunits duplicated from the natural vegetated landunit in the CLM sub-grid hierarchy make it possible to prescribe up to three different excess ice conditions in each grid cell with permafrost.

484 A test over the Lena river delta showcases that the sub-grid representation of excess ice can retrieve 485 the sub-grid variability of annual thaw-freeze state and the excess ice melt and surface subsidence 486 through time. On the other hand, initializing excess ice homogeneously throughout the grid cell produces 487 a smaller stabilization effect of excess ice to the permafrost thermal regime and the local surface 488 subsidence under a warming climate. With a tiling scheme ingesting a global data set of excess ice 489 condition into the CLM surface data, our model development shows the capability of portraying more details on simulating permafrost degradation trajectories. As excess ice thermally stabilizes the 490 491 permafrost on the sub-grid scale, permafrost degrades with a trajectory from continuous permafrost to 492 discontinuous permafrost, and finally to a permafrost-free area. The modeled global pattern of permafrost therefore exhibits regions of discontinuous permafrost as the transition zone between the continuouspermafrost and degraded permafrost.

495 This study, for the first time, used an ESM to project excess ice melt and surface subsidence and 496 permafrost degradation with sub-grid variability. The approach of duplicating tiles at the landunit level 497 instead of the column level allows more freedom for further developments in this direction. Furthermore, 498 the new CLM tiling hierarchy has much more potential than representing more accurate excess ice 499 physics as examined in this study. The accuracies of predicted excess ice melt and surface subsidence 500 trends are limited at present by the available global-scale dataset and studies on excess ground ice 501 conditions, thus further advancement of the excess ice modeling will rely on new or improved 502 observational studies or datasets of the excess ground ice conditions at the global scale. The model 503 development in our study, therefore, lays the foundation for further advances focusing on excess ice 504 modeling and other processes in the CLM framework that could benefit from an improved sub-grid 505 representation.

506

507 Source code and data availability

have improved this paper substantially.

The original Community Land Model is available at <u>https://github.com/ESCOMP/ctsm</u>. The source code
 of model development in this study is available from the corresponding author upon request.

510 Author contributions

L.C conducted model development work and wrote the initial draft with additional contributions from
all authors. H.L, S.W, and K.S.A provided ideas and help during the process of model development. H.L
provided the code of excess ice physics in the earlier version of CLM. L.C prepared all figures.

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Table 1: The excess ice initialization scenario in each of the three terraces (landunits) for the Lena River delta, as well as that for the single-landunit excess ice initialization case.

/21	River delta, as well as that for the single-landunit excess ice initialization c
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Depth (after adding ice)	Volumetric Ice content	Area weight					
No excess ice terrain							
N/A	0%	24.6%					
Holocene ground ice terrain							
0.9-9 m	65%	66.6%					
	Yedoma ice complex						
0.6-20 m	90%	8.8%					
А	verage ice single-landunit case						
0.6-0.9 m	7.92%	100%					
0.9-9 m	51.21%	100%					
9-20 m	7.92%	100%					

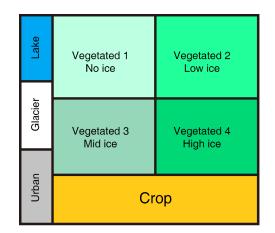
Overall visible ground ice content for each CAPS point	Tiling scheme (area weights for each excess ice category)	Eligible CAPS types	
5%	80% no excess ice; 20% Low	clf; clf; slf; ilf; clr; dlr; slr; ilr	
15%	58% no excess ice; 20% Low; 22% Medium	cmf; dmf; smf; imf; dhr; shr; ihr	
15%	66% no excess ice; 20% Low; 14% High	chr	
25%	44% no excess ice; 20% Low; 22% Medium; 14% High	dhf; shf; ihf	
25%	52% no excess ice; 20% Low; 28% High	chf	

Table 2: The tiling scheme prescribing area weights of landunits for each CAPS class. The detailed CAPS classes are shown in Figure 2.

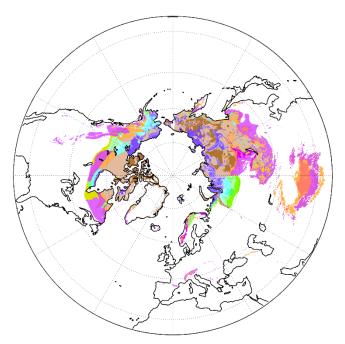
727 Note: For each class, the first letter is for the permafrost extent, the second for the excess ice content, and the third
728 for the terrain and overburden, following Brown et al. (2002).

730 Table 3: List of simulations conducted for this study.

Cases	Description						
Single point cases for the Lena river delta							
Triple-landunit case	Applying the sub-grid representation of excess ice. Three natural vegetated landunit initialized.						
Average ice single- landunit case	Not applying the sub-grid representation of excess ice. Only one natural vegetated landunit initialized. The grid-mean excess ice content for each soil layer in the only landunit is calculated by spatially averaging those in different landunits in the triple-landunit case.						
	Global simulation cases						
No ice case	Not adding any excess ground ice (the original CLM5 simulation).						
Sub-grid ice case	Applying the sub-grid representation of excess ice. A tiling scheme helps to "translate" excess ice conditions in the CAPS data to fit what the CLM5 requires.						
Grid-average ice case	Not applying the sub-grid representation of excess ice. The grid-mean excess ice content for each soil layer is calculated by spatially averaging those in different landunits in the sub-grid ice case.						



- 734 Figure 1: Modification of the CLM5 tiling hierarchy on the landunit level containing four natural
- 735 vegetated landunits for different excess ice conditions.



Permafrost area classification

	Ground Ice Content (percent by volume)						
Permafrost Extent	Lowlands, highlands, and intra-and intermontane depressions			Mountains, highlands, ridges, and plateaus			
	25%	15%	5%	15%	5%		
Continous (100%)	chf	cmf	clf	chr	clr		
Discontinous (70%)	dhf	dmf	dlf	dhr	dlr		
Sporadic (30%)	shf	smf	slf	shr	slr		
Isolated (5%)	ihf	imf	ilf	ihr	ilr		
* Letter code naming: The first letter is for the permafrost extent, second for the ground excess ice concent, and the thrid for the terrain and overburden.							

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Figure 2: Spatial distribution of excess ground ice in the Northern Hemisphere modified from
Brown et al. (2002). Compared to the original data, permafrost extents and ground ice contents
are converted to definite numbers (percentages) for model computation.

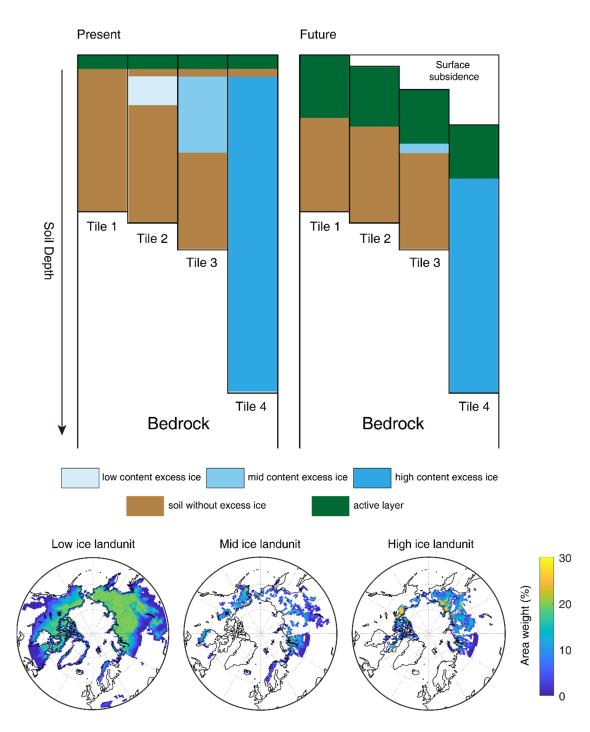
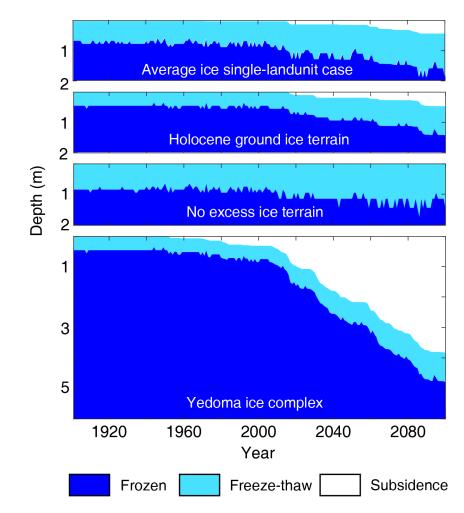


Figure 3. Schematic representation of the sub-grid excess ice initialization scenario, and maps
showing the area weight (%) occupied by different excess ice landunits, i.e. the initial condition of
excess ice in the global simulation.



749 Figure 4. Annual freeze-thaw state for the three terraces for the triple-landunit case, as well as for

750 the average ice single-landunit case.

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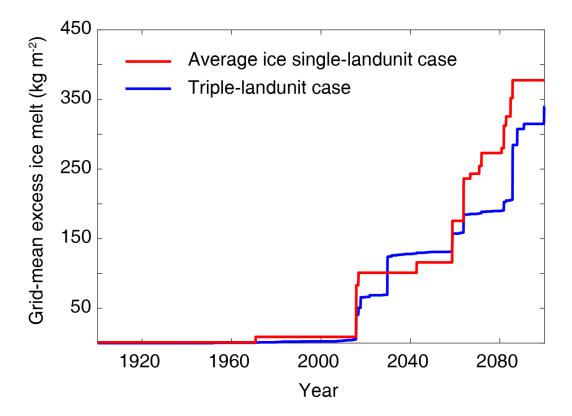


Figure 5. Grid-mean excess ice melt since 1900 for the single-point cases over the Lena river delta
with and without the sub-grid excess ice initialization.

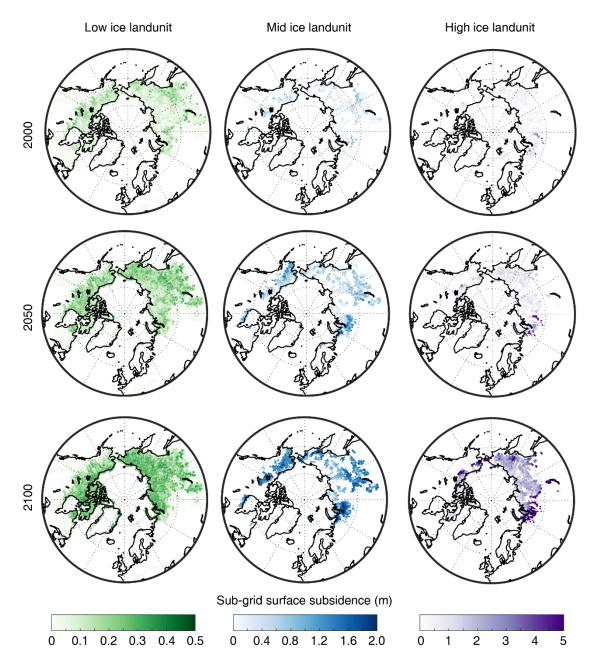


Figure 6. Maps showing sub-grid surface subsidence (m) in 2000, 2050, 2100 in the low, mid, and
high excess ice landunits in the sub-grid ice case.

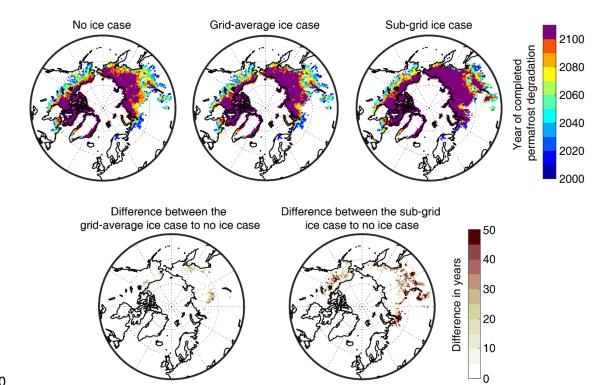
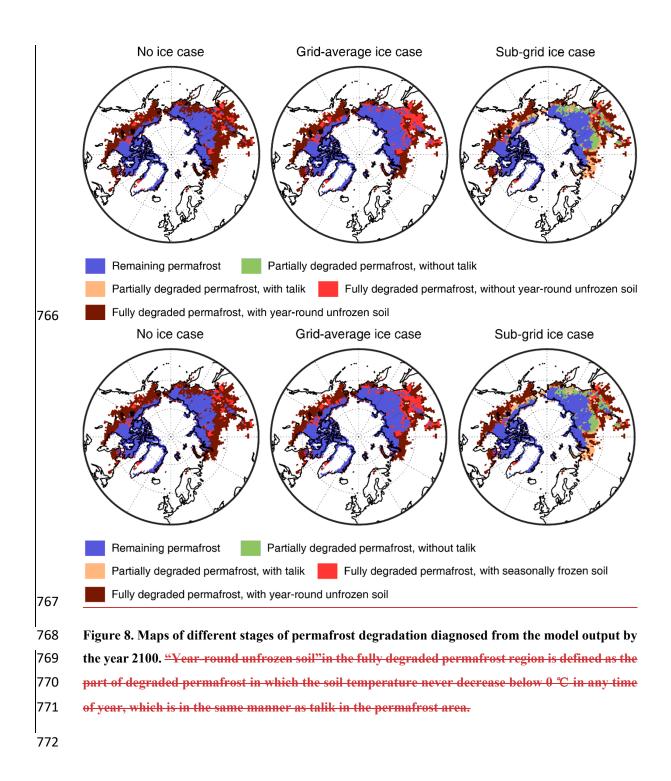


Figure 7. Maps showing the year of completed permafrost degradation (upper set of three maps),
as well as the differences between cases (lower set of two maps). The purple color indicates the
existence of permafrost in these grid cells by 2100. The difference in years is provided only for grid
cell with completed permafrost degradation before 2100.



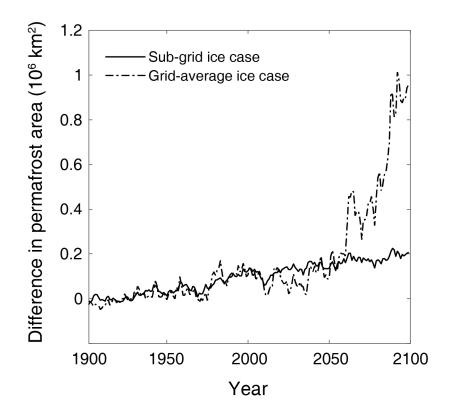




Figure 9. Difference in modeled permafrost area versus time between the sub-grid ice case and no
ice case, as well as between the grid-average ice case and no ice case.